

```
In [1]: # Analisis Anova multifactorial
```

```
# 1. Carga inicial de datos.
```

```
if(!require(psych)){install.packages("psych")}
if(!require(FSA)){install.packages("FSA")}
if(!require(ggplot2)){install.packages("ggplot2")}
if(!require(car)){install.packages("car")}
if(!require(multcompView)){install.packages("multcompView")}
if(!require(lsmmeans)){install.packages("lsmmeans")}
if(!require(rcompanion)){install.packages("rcompanion")}
```

```
ln <- ("Algoritmo      Entrenamiento      Rendimiento
'Algoritmo A'      MT500      12000
'Algoritmo A'      MT500      14005
'Algoritmo A'      MT500      13508
'Algoritmo A'      MT500      9503
'Algoritmo A'      MT500      14004
'Algoritmo A'      MT1000     11502
'Algoritmo A'      MT1000     13006
'Algoritmo A'      MT1000     13252
'Algoritmo A'      MT1000     14253
'Algoritmo A'      MT1000     15003
'Algoritmo A'      MT5000     12504
'Algoritmo A'      MT5000     11504
'Algoritmo A'      MT5000     9500
'Algoritmo A'      MT5000     11506
'Algoritmo A'      MT5000     16000
'Algoritmo A'      MT50000    13008
'Algoritmo A'      MT50000    10506
'Algoritmo A'      MT50000    13005
'Algoritmo A'      MT50000    17002
'Algoritmo A'      MT50000    13008
'Algoritmo B'      MT500      11005
'Algoritmo B'      MT500      12007
'Algoritmo B'      MT500      12509
'Algoritmo B'      MT500      10504
'Algoritmo B'      MT500      12002
'Algoritmo B'      MT1000     12504
'Algoritmo B'      MT1000     13501
'Algoritmo B'      MT1000     13501
'Algoritmo B'      MT1000     13252
'Algoritmo B'      MT1000     15256
'Algoritmo B'      MT5000     12253
'Algoritmo B'      MT5000     11255
'Algoritmo B'      MT5000     10006
'Algoritmo B'      MT5000     11252
'Algoritmo B'      MT5000     14004
'Algoritmo B'      MT50000    12007
'Algoritmo B'      MT50000    11505
'Algoritmo B'      MT50000    14009
'Algoritmo B'      MT50000    15000
'Algoritmo B'      MT50000    12009
'Algoritmo C'      MT500      9000
'Algoritmo C'      MT500      11003
'Algoritmo C'      MT500      11505
'Algoritmo C'      MT500      9509
'Algoritmo C'      MT500      11003
'Algoritmo C'      MT1000     11508
'Algoritmo C'      MT1000     12508
'Algoritmo C'      MT1000     12506
'Algoritmo C'      MT1000     12254
'Algoritmo C'      MT1000     13253
'Algoritmo C'      MT5000     11255
'Algoritmo C'      MT5000     10257
'Algoritmo C'      MT5000     9500
'Algoritmo C'      MT5000     9255
'Algoritmo C'      MT5000     12009
'Algoritmo C'      MT50000    11000
'Algoritmo C'      MT50000    9509
'Algoritmo C'      MT50000    13009
'Algoritmo C'      MT50000    14005
'Algoritmo C'      MT50000    11001
")
```

```

# Se introduce la tabla.
Data <- read.table(textConnection(ln), header = TRUE)

# Se ordenan los datos según los ingresamos. (Evitar orden alfabético por R).
Data$Entrenamiento <- factor(Data$Entrenamiento, levels=unique(Data$Entrenamiento))

# 2. Verificación de la lectura de datos

library(psych)
headTail(Data)
str(Data)
summary(Data)
rm(ln)

# 3. Gráfico simple de interacción.

interaction.plot(x.factor = Data$Entrenamiento,
  trace.factor = Data$Algoritmo,
  response = Data$Rendimiento,
  fun = mean,
  type = "b",
  col = c("black", "red", "green"),
  pch = c(19,17,15),
  fixed = TRUE,
  leg.bty = "o")

# 4. Modelo lineal y ANOVA

model <- lm(Rendimiento ~ Entrenamiento + Algoritmo + Entrenamiento : Algoritmo, data = Data)

library(car)
Anova(model, type = "II")

# 5. Evaluación de supuestos

# Normalidad
x <- residuals(model)
library(rcompanion)
plotNormalHistogram(x)

# Dispersión de los residuos
plot(fitted(model), residuals(model))

# Graficos del modelo lineal
plot(model)

# 6. Análisis post-hoc

library(lsmeans)
marginal <- lsmeans(model, pairwise ~ Algoritmo, adjust = "tukey")
marginal

# Funcion cld
library(multcomp)
CLD <- cld(marginal, alpha=0.05, Letters= letters, adjust="tukey")
CLD

# Análisis post-hoc entrenamiento
marginal <- lsmeans(model, pairwise ~ Entrenamiento, adjust = "tukey")
marginal

# Funcion cld
library(multcomp)
CLD <- cld(marginal, alpha=0.05, Letters= letters, adjust="tukey")
CLD

# 7. Gráfico final
library(FSA)

Sum <- Summarize(Rendimiento ~ Entrenamiento + Algoritmo, data = Data, digits = 3)
Sum$se <- Sum$sd / sqrt(Sum$n)
Sum$se <- signif(Sum$se, digits = 3)
Sum

```

```

Sum$Entrenamiento <- factor(Sum$Entrenamiento,
  levels = unique(Sum$Entrenamiento))

# 8. Boxplot error estándar

library(FSA)
library(ggplot2)
pd <- position_dodge(.2)
ggplot(Sum, aes(x=Entrenamiento,
  y = mean,
  color = Algoritmo)) +
  geom_errorbar(aes(ymin=mean-se,
    ymax=mean + se),
  width=.2, size=0.7, position=pd) +
  geom_point(shape=15, size=4, position = pd) +
  theme_bw() +
  theme(axis.title = element_text(face="bold")) +
  scale_colour_manual(values = c("black", "red", "green")) +
  ylab("Rendimiento")

```

Loading required package: psych

Loading required package: FSA

FSA v0.9.4. See citation('FSA') if used in publication.

Run fishR() for related website and fishR('IFAR') for related book.

Attaching package: 'FSA'

The following object is masked from 'package:psych':

headtail

Loading required package: ggplot2

Attaching package: 'ggplot2'

The following objects are masked from 'package:psych':

%+%, alpha

Loading required package: car

Loading required package: carData

Registered S3 methods overwritten by 'car':

method from

hist.boot FSA

confint.boot FSA

Attaching package: 'car'

The following object is masked from 'package:FSA':

bootCase

The following object is masked from 'package:psych':

logit

Loading required package: multcompView

Loading required package: lsmeans

Loading required package: emmeans

The 'lsmeans' package is now basically a front end for 'emmeans'.
Users are encouraged to switch the rest of the way.

See help('transition') for more information, including how to
convert old 'lsmeans' objects and scripts to work with 'emmeans'.

Loading required package: rcompanion

Attaching package: 'rcompanion'

The following object is masked from 'package:psych':

phi

A data.frame: 9 × 3

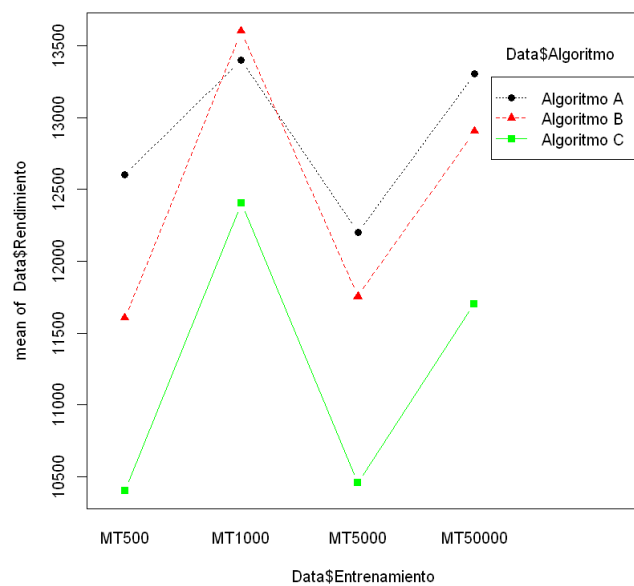
	Algoritmo	Entrenamiento	Rendimiento
	<chr>	<fct>	<chr>
1	Algoritmo A	MT500	12000
2	Algoritmo A	MT500	14005
3	Algoritmo A	MT500	13508
4	Algoritmo A	MT500	9503
...	NA	NA	...
57	Algoritmo C	MT50000	9509
58	Algoritmo C	MT50000	13009
59	Algoritmo C	MT50000	14005
60	Algoritmo C	MT50000	11001

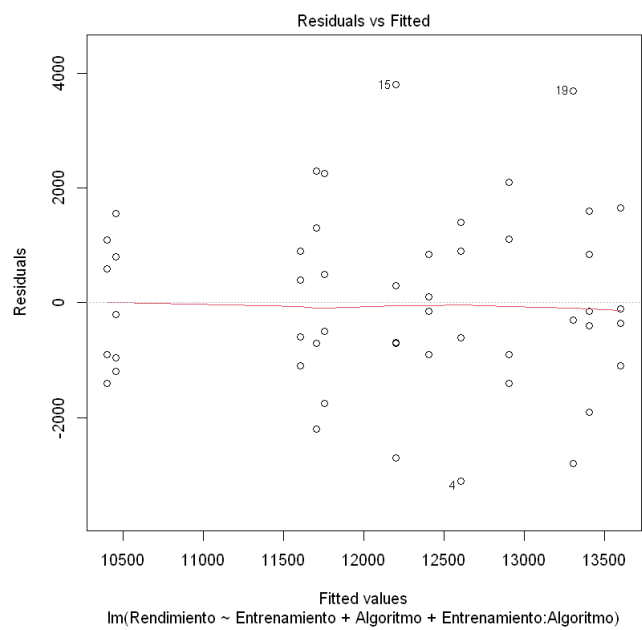
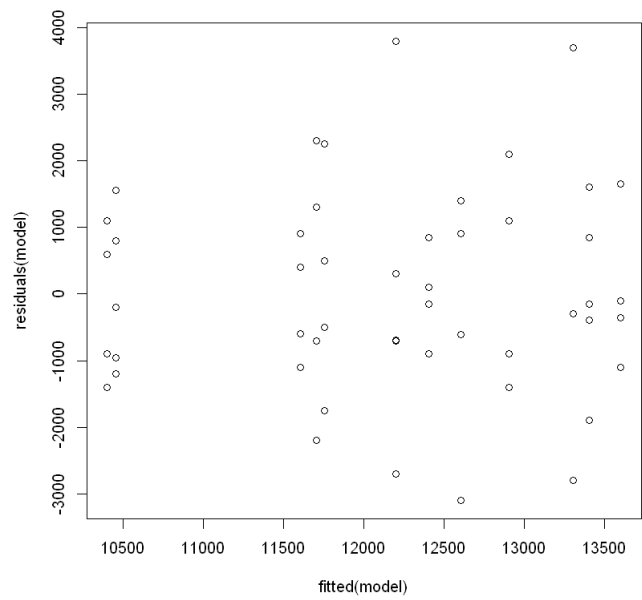
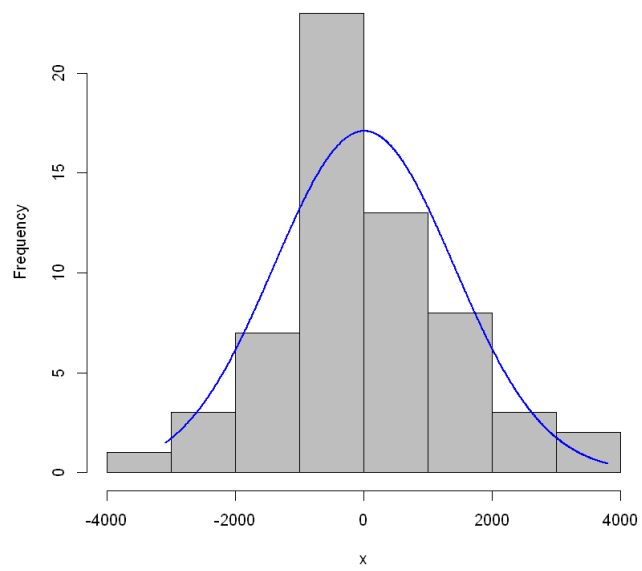
```
'data.frame': 60 obs. of 3 variables:
 $ Algoritmo : chr "Algoritmo A" "Algoritmo A" "Algoritmo A" "Algoritmo A" ...
 $ Entrenamiento: Factor w/ 4 levels "MT500","MT1000",...: 1 1 1 1 1 2 2 2 2 2 ...
 $ Rendimiento : int 12000 14005 13508 9503 14004 11502 13006 13252 14253 15003 ...

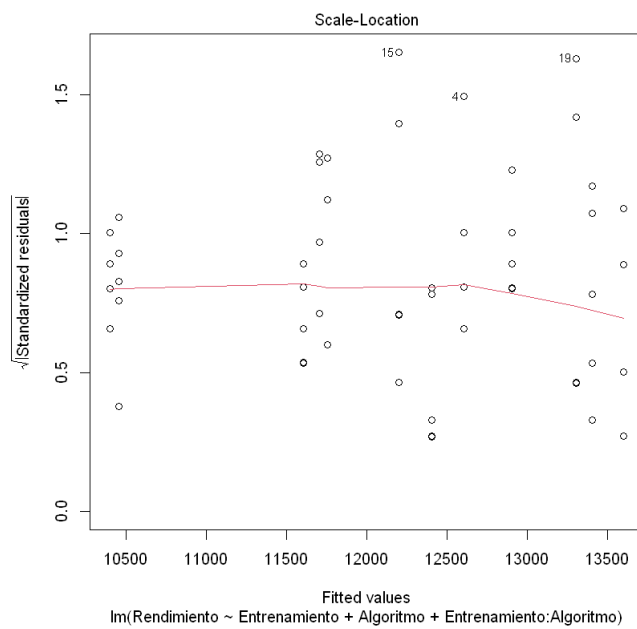
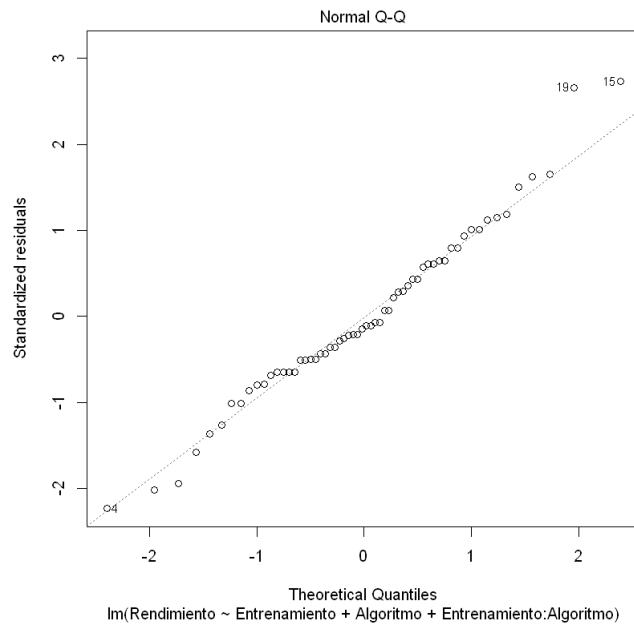
  Algoritmo      Entrenamiento  Rendimiento
Length:60      MT500 :15      Min. : 9000
Class :character MT1000 :15     1st Qu.:11004
Mode :character  MT5000 :15     Median :12009
                  MT50000:15    Mean :12196
                  Max. :17002
```

A anova: 4 × 4

	Sum Sq	Df	F value	Pr(>F)
	<dbl>	<dbl>	<dbl>	<dbl>
Entrenamiento	30621741	3	4.2341389	0.009822338
Algoritmo	28982927	2	6.0113044	0.004679860
Entrenamiento:Algoritmo	2444325	6	0.1689912	0.983866401
Residuals	115713695	48	NA	NA







NOTE: Results may be misleading due to involvement in interactions

\$lsmeans

\$bhat

\$V

\$levels

\$linfct

\$dffun

\$dfargs

\$post.beta

\$estName

\$estType

\$infer

\$level

\$adjust

\$famSize

\$avgd.over

\$sigma

\$methDesc

\$extras

\$contrasts

\$bhat

.wgt.
<dbl>
20
20
20

\$V

\$levels

\$linfct

\$dffun

\$dfargs
\$post.beta

\$estName
\$estType
\$infer

\$level
\$adjust
\$famSize
\$avgd.over
\$methDesc

	(Intercept) En
(Intercept)	482140.4
EntrenamientoMT1000	-482140.4
EntrenamientoMT5000	-482140.4
EntrenamientoMT50000	-482140.4
AlgoritmoAlgoritmo B	-482140.4
AlgoritmoAlgoritmo C	-482140.4
EntrenamientoMT1000:AlgoritmoAlgoritmo B	482140.4
EntrenamientoMT5000:AlgoritmoAlgoritmo B	482140.4
EntrenamientoMT50000:AlgoritmoAlgoritmo B	482140.4
EntrenamientoMT1000:AlgoritmoAlgoritmo C	482140.4
EntrenamientoMT5000:AlgoritmoAlgoritmo C	482140.4
EntrenamientoMT50000:AlgoritmoAlgoritmo C	482140.4

\$contrast =
'Algoritmo A - Algoritmo B' · 'Algoritmo A - Algoritmo C' ·

(Intercept)	EntrenamientoMT1000	EntrenamientoMT5000
0	0	0
0	0	0
0	0	0

function (k, dfargs)
dfargs\$df
\$df = 48
A
matrix:
1 × 1
of
type
lgl
NA

'estimate'
'pairs'
FALSE · TRUE

0.95
'tukey'
3
'Entrenamiento'
'pairwise differences'

`$is.new.rg`
`$.pairby`
`$orig.grid`

FALSE
"
A data.frame:
3 × 1
Algoritmo
<fct>
Algoritmo A
Algoritmo B
Algoritmo C

`$con.coef`

A matrix: 3 × 3 of type dbl

	Algoritmo A	Algoritmo B	Algoritmo C
Algoritmo A - Algoritmo B	1	-1	
Algoritmo A - Algoritmo C	1	0	
Algoritmo B - Algoritmo C	0	1	

Loading required package: mvtnorm
Loading required package: survival
Loading required package: TH.data
Loading required package: MASS

Attaching package: 'TH.data'

The following object is masked from 'package:MASS':

geyser

Note: adjust = "tukey" was changed to "sidak"
because "tukey" is only appropriate for one set of pairwise comparisons

A summary_emm: 3 × 7

	Algoritmo	lsmean	SE	df	lower.CL	upper.CL	.group
	<fct>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<chr>
3	Algoritmo C	11242.45	347.1817	48	10383.55	12101.35	a
2	Algoritmo B	12467.05	347.1817	48	11608.15	13325.95	b
1	Algoritmo A	12878.95	347.1817	48	12020.05	13737.85	b

NOTE: Results may be misleading due to involvement in interactions

\$lsmmeans

\$bhat

\$V

\$levels

\$linfct

\$dffun

\$dfargs

\$post.beta

\$estName

\$estType

\$infer

\$level

\$adjust

\$famSize

\$avgd.over

\$sigma

\$methDesc

\$extras

.wgt.
<dbl>
15
15
15
15

\$bhat

\$V

12604 · 799.199999999997 · -401.200000000003 · 701.7995

	(Intercept) En
(Intercept)	482140.4
EntrenamientoMT1000	-482140.4
EntrenamientoMT5000	-482140.4
EntrenamientoMT50000	-482140.4
AlgoritmoAlgoritmo B	-482140.4
AlgoritmoAlgoritmo C	-482140.4
EntrenamientoMT1000:AlgoritmoAlgoritmo B	482140.4
EntrenamientoMT5000:AlgoritmoAlgoritmo B	482140.4
EntrenamientoMT50000:AlgoritmoAlgoritmo B	482140.4
EntrenamientoMT1000:AlgoritmoAlgoritmo C	482140.4
EntrenamientoMT5000:AlgoritmoAlgoritmo C	482140.4
EntrenamientoMT50000:AlgoritmoAlgoritmo C	482140.4

\$levels

\$contrast =
'MT500 - MT1000' · 'MT500 - MT5000' · 'MT500 - MT5000C

\$linfct

(Intercept)	EntrenamientoMT1000	EntrenamientoMT5000
0	-1	0
0	0	-1
0	0	0
0	1	-1
0	1	0
0	0	1

\$dffun

function (k, dfargs)

\$dfargs

dfargs\$df

\$post.beta

\$df = 48

A
matrix:
1 × 1
of
type
lgl
NA

\$estName

'estimate'

\$estType

'pairs'

\$infer	FALSE · TRUE
\$level	0.95
\$adjust	'tukey'
\$famSize	4
\$avgd.over	'Algoritmo'
\$methDesc	'pairwise differences'
\$is.new.rg	FALSE
\$.pairby	"
\$orig.grid	A data.frame: 4 × 1
	Entrenamiento
	<fct>
	MT500
	MT1000
	MT5000
	MT50000
\$con.coef	A matrix: 6 × 4 of type dbl
	MT500 MT1000 MT5000 MT50000
	MT500 - MT1000 1 -1 0 0
	MT500 - MT5000 1 0 -1 0
	MT500 - MT50000 1 0 0 -1
	MT1000 - MT5000 0 1 -1 0
	MT1000 - MT50000 0 1 0 -1
	MT5000 - MT50000 0 0 1 -1

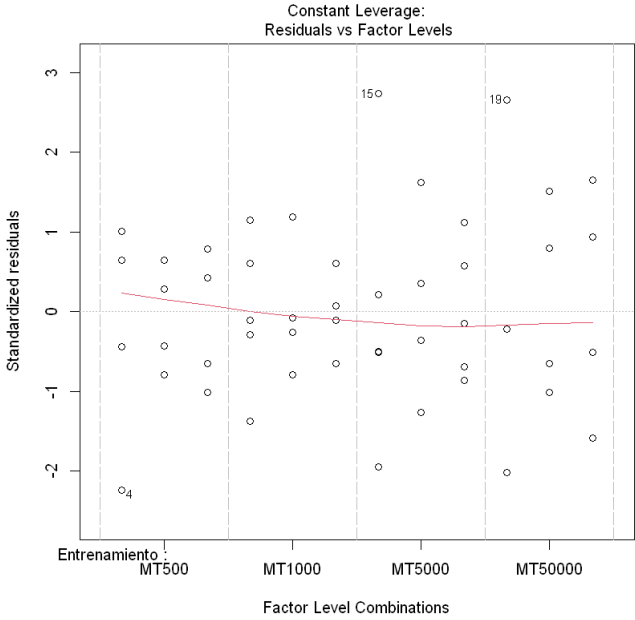
Note: adjust = "tukey" was changed to "sidak" because "tukey" is only appropriate for one set of pairwise comparisons

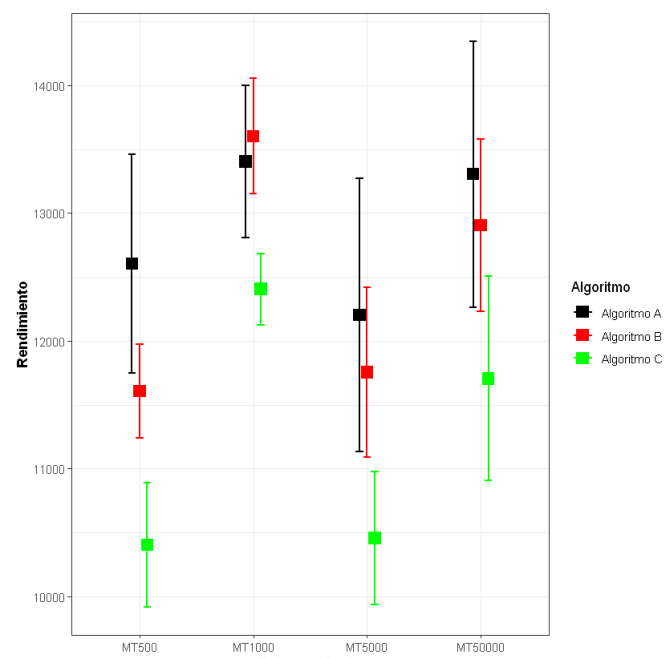
A summary_emm: 4 × 7							
	Entrenamiento	lsmean	SE	df	lower.CL	upper.CL	.group
	<fct>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<chr>
3	MT5000	11470.67	400.8908	48	10433.24	12508.09	a
1	MT500	11537.80	400.8908	48	10500.38	12575.22	a
4	MT50000	12638.87	400.8908	48	11601.44	13676.29	ab
2	MT1000	13137.27	400.8908	48	12099.84	14174.69	b

A data.frame: 12 × 11

Entrenamiento	Algoritmo	n	mean	sd	min	Q1	median	Q3	max	se
<fct>	<chr>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>
MT500	Algoritmo A	5	12604.0	1918.400	9503	12000	13508	14004	14005	858
MT1000	Algoritmo A	5	13403.2	1330.104	11502	13006	13252	14253	15003	595
MT5000	Algoritmo A	5	12202.8	2386.861	9500	11504	11506	12504	16000	1070
MT50000	Algoritmo A	5	13305.8	2332.843	10506	13005	13008	13008	17002	1040
MT500	Algoritmo B	5	11605.4	822.896	10504	11005	12002	12007	12509	368
MT1000	Algoritmo B	5	13602.8	1010.473	12504	13252	13501	13501	15256	452
MT5000	Algoritmo B	5	11754.0	1488.960	10006	11252	11255	12253	14004	666
MT50000	Algoritmo B	5	12906.0	1514.683	11505	12007	12009	14009	15000	677
MT500	Algoritmo C	5	10404.0	1084.210	9000	9509	11003	11003	11505	485
MT1000	Algoritmo C	5	12405.8	625.963	11508	12254	12506	12508	13253	280
MT5000	Algoritmo C	5	10455.2	1166.849	9255	9500	10257	11255	12009	522
MT50000	Algoritmo C	5	11704.8	1789.237	9509	11000	11001	13009	14005	800

Warning message:
"Using `size` aesthetic for lines was deprecated in ggplot2 3.4.0.
Please use `linewidth` instead."





In []: